ZILFIMIAN



Regularization/R (Q7L8)

30% (6/20)

1. In Poisson regression...

- (A) The asymptotic distribution of the maximum likelihood estimates is multivariate normal.
- (B) The distribution of the maximum likelihood estimates is multivariate normal.
- The asymptotic distribution of the maximum likelihood estimates is multivariate Poisson distribution.
- D I do not know

✓ 2. In the case of intercept-only model

- A The mean of the dependent variable equals the exponential value of intercept
- B The mean of the dependent variable equals the intercept
- (c) The mean of the dependent variable equals 0
- D I do not know

3. In(lambda) = 0.6 - 0.2* female [lamda = the average number of articles] Note: e^(-0.2)=0.78

- A One unit increase in female brings a 0.2 decrease in ln(lambda).
- (B) Being female decreases the average number of articles by 0.78 percent
- (c) Being female decreases the average number of articles by 22%
- D I do not know

× 4. In the multiple linear regression, we assume that...

- (A) The number of observations is much larger than the number of variables (n>>p)
- (B) The number of observations is slightly larger than the number of variables (n>p)
- (C) The number of observations equals than the number of variables (n=p)
- (n<p) The number of observations is lees than the number of variables (n<p)
- E It is not important
- (F) I do not know

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| × | 5. | The way of solving the problem of a large number of variables is Subset Selection & Shrinkage (Regularization) |
|---|----------------|---|
| | (B) | Shrinkage (Regularization) & Maximum Likelihood estimation |
| | \overline{C} | Dimension Reduction & OLS estimation |
| | | I do not know |
| | E | The absence of the right answer |
| × | 6. (A) | The bias of an estimator (e.g. z^{-}) equalsHint: the OLS coefficients are unbias :) $E(z^{-}) - z$ |
| | B | $E(z^2) - [E(z)]^2$ |
| | (C) | $[E(z^2) - E(z)]^2$ |
| | D | E(z^2) |
| | E | I do not know |
| × | | The main idea of regularization is |
| | A | To introduce a small amount of bias in order to have less variance. |
| | (B) | To introduce a small amount of variance in order to have less bias. |
| | (c) | To introduce a small amount of variance and bias in order to have less bias. |
| | (D) | I do not know |
| × | 8. | With which function we can show regularization in R |
| | (A) | glmnet() |
| | (B) | regular() |
| | C | lm() |
| | D | glm() |
| | E | I do not know |
| × | 9. | How the tune of any parametr can be made |
| | (A) | using Cross validation |
| | B | It is impossible |
| | \bigcirc | I do not now |
| | D | using larger sample |
| | E | only having population |

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the combination of L1 and L2 regularization the combination of L2 and L3 regularization is independent from other types of refularization I do not know not a type of regularization **X** 11. Regularization is used only for Poisson Regression Linear Regression Logistic Regression any regression I do not know Regularization can solve the problem of 12. heteroscedasticity multicollinearity autocorrelation I do not know As a result of regularization we will have smaller slope than in case of OLS larger slope than in case of OLS the slope remains the same I do not know **X** 14. The ridge coefficient estimates shrink towards zero when lambda increases when lambda decreases when lambda = 0 I do not know

X 10.

Elastic Net is

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| / | | Which one can shrink the slope all the way to 0? |
|----------|-------------|--|
| | | Lasso |
| | \sim | Ridge |
| | \sim | Regression |
| | (D) | I do not know |
| / | 16. | When lambda = 0, we have |
| | (A) | Ridge |
| | B | Lasso |
| | C | EL |
| | D | Regression |
| | E | I do not know |
| × | 17 | When alpha = 0, we have |
| • | | Ridge |
| | \simeq | Lasso |
| | \bigcirc | EL . |
| | | Regression |
| | \sim | I do not know |
| | | |
| × | 18. knov | variables need to be incorporated in the model according to domain wledge |
| | This | statement is true for |
| | (A) | Ridge |
| | B | Lasso |
| | C | EL |
| | D | Regression |
| | E | I do not know |
| / | 19. | Which function can help to perform cross-validation for regularization in R? |
| | A | cv.glmnet() |
| | B | cros_val() |
| | (c) | glmnet(method = "cv) |
| | | I do not know |
| | | |

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20. Why we use set.seed() in R?



To have universal result



(B) To perform better result



C To have random models

I do not know

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